

# Business Presentation

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# Business Problem Overview and Solution Approach

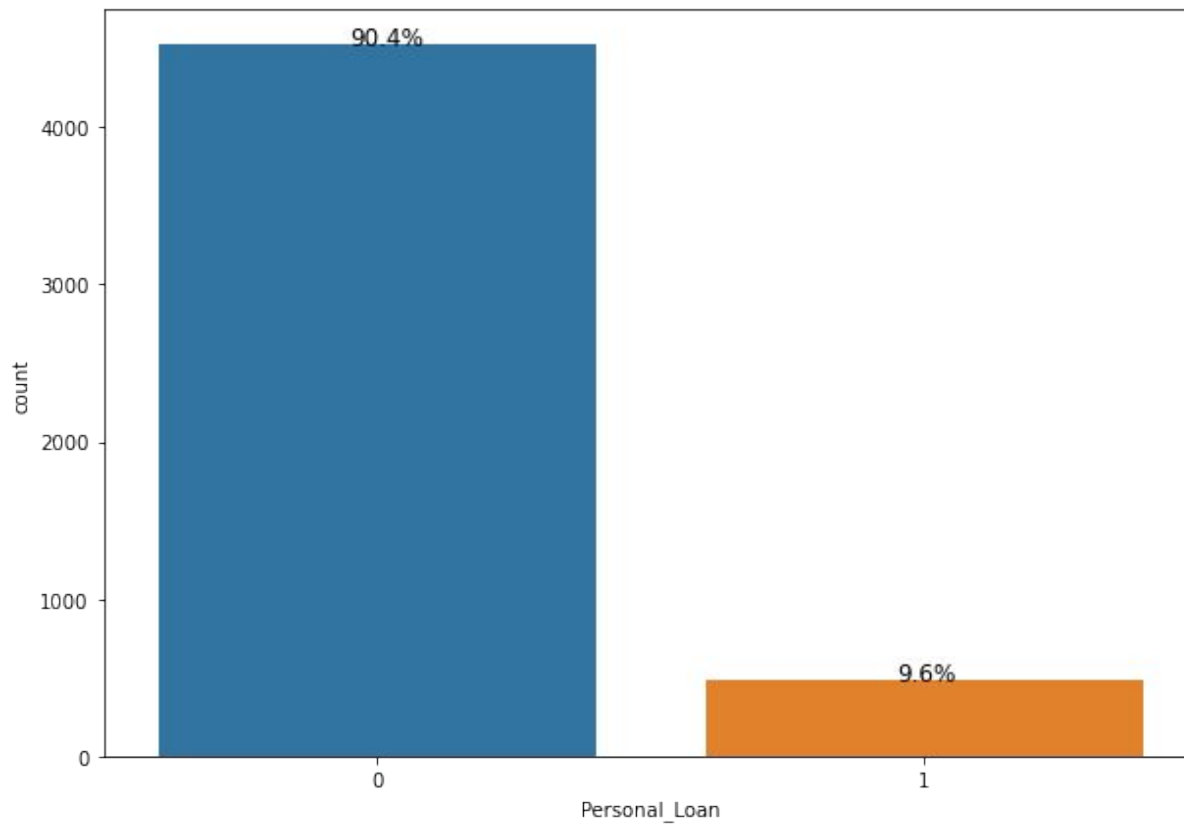
- Core business idea
  - A campaign that the bank ran last year for liability customers showed a healthy conversion rate of over 9% success. This has encouraged the retail marketing department to devise campaigns with better target marketing to increase the success ratio.
- Problem to tackle
  - Build a model that will help the marketing department to identify the potential customers who have a higher probability of purchasing the loan.

# Data Overview

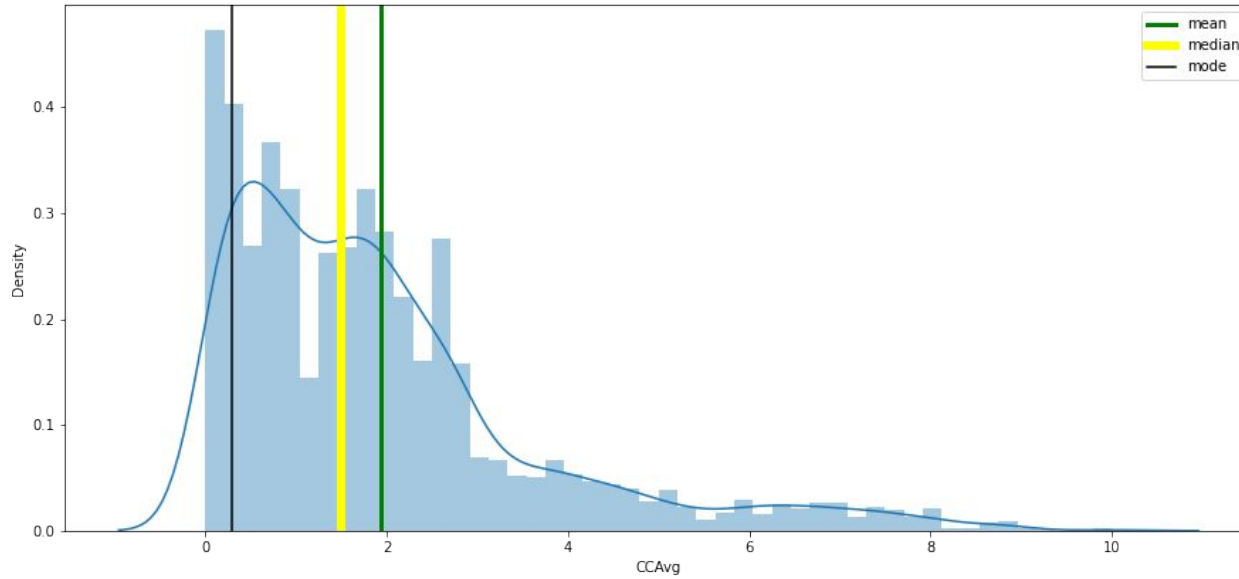
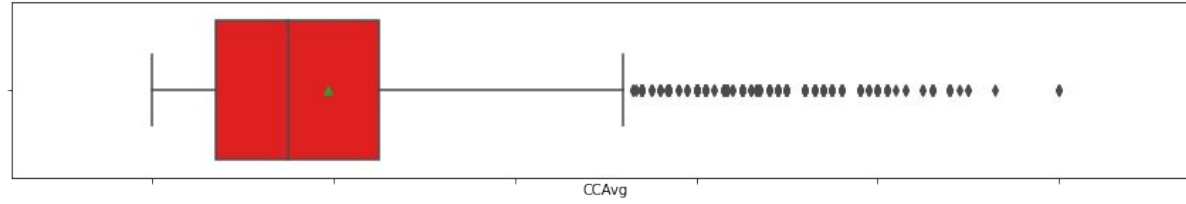
- **Data Dictionary**

- ID: Customer ID
- Age: Customer's age in completed years
- Experience: #years of professional experience
- Income: Annual income of the customer (in thousand dollars)
- ZIP Code: Home Address ZIP code.
- Family: the Family size of the customer
- CCAvg: Average spending on credit cards per month (in thousand dollars)
- Education: Education Level. 1: Undergrad; 2: Graduate; 3: Advanced/Professional
- Mortgage: Value of house mortgage if any. (in thousand dollars)
- Personal\_Loan: Did this customer accept the personal loan offered in the last campaign?
- Securities\_Account: Does the customer have securities account with the bank?
- CD\_Account: Does the customer have a certificate of deposit (CD) account with the bank?
- Online: Do customers use internet banking facilities?
- CreditCard: Does the customer use a credit card issued by any other Bank (excluding All life Bank)

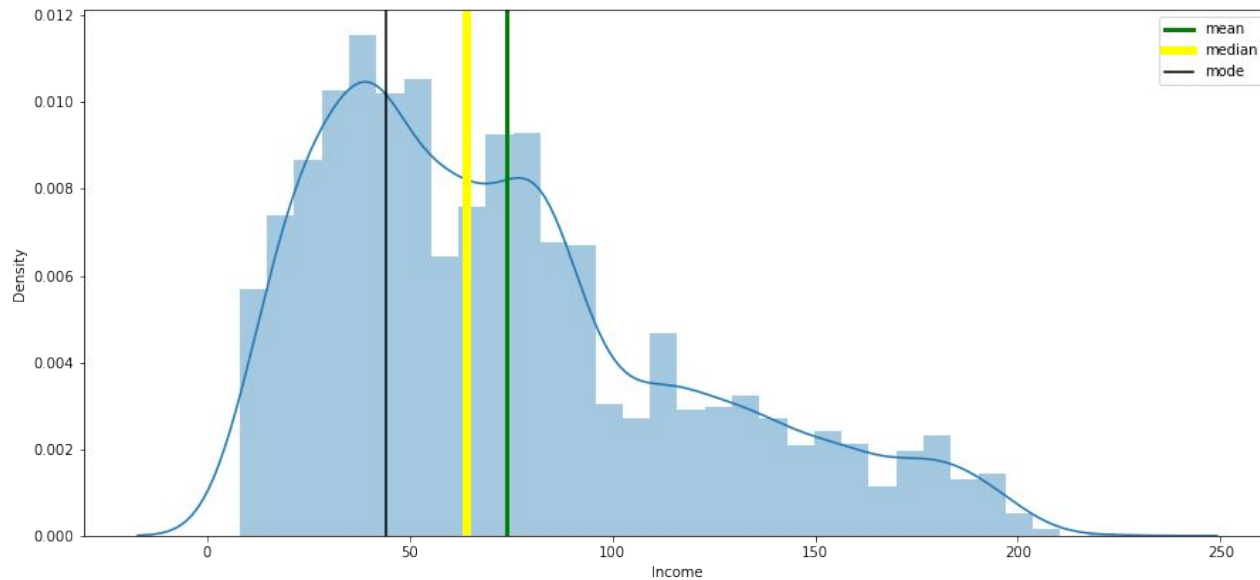
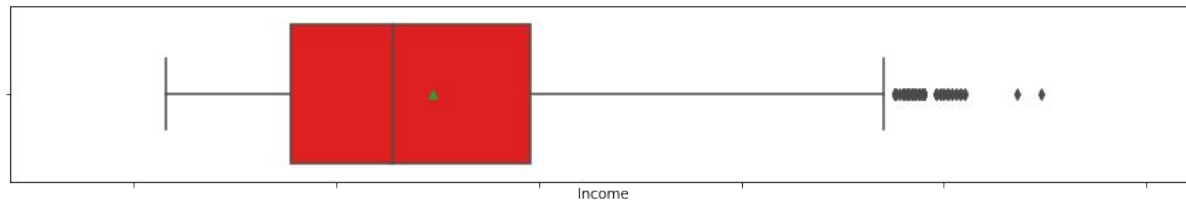
## EDA



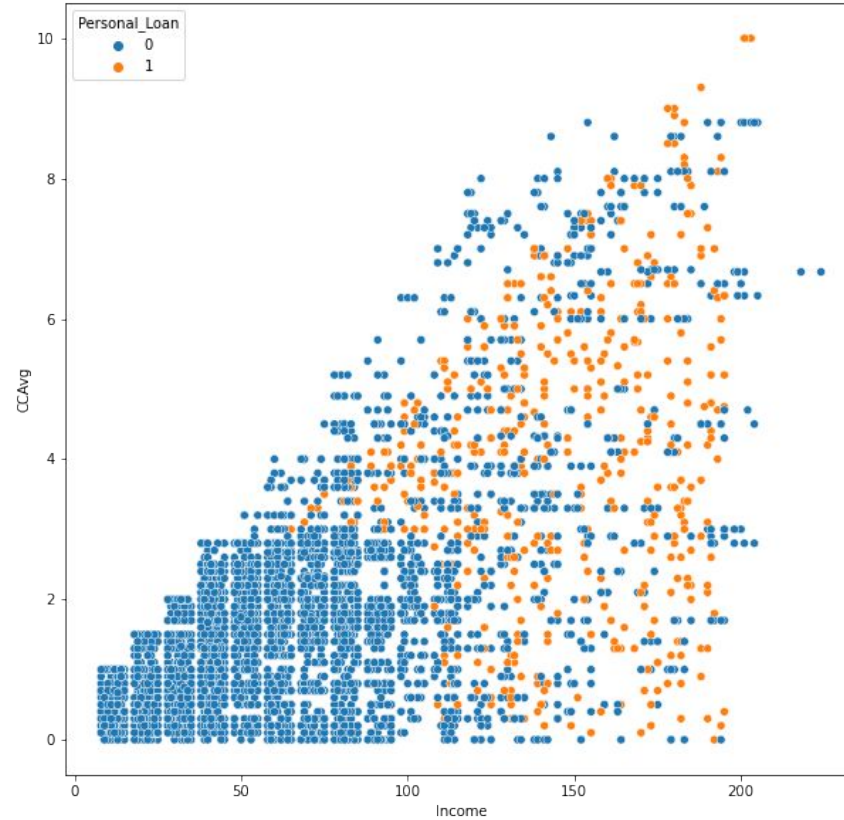
# EDA



# EDA

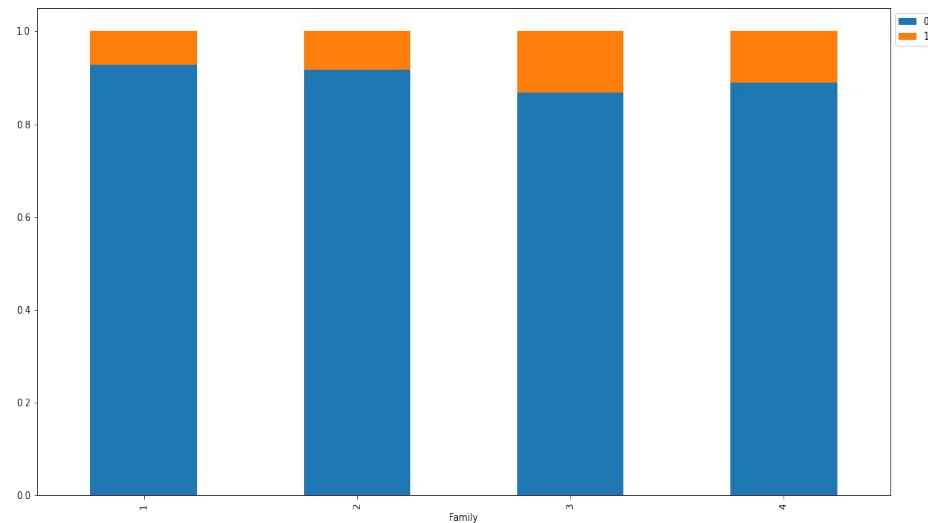
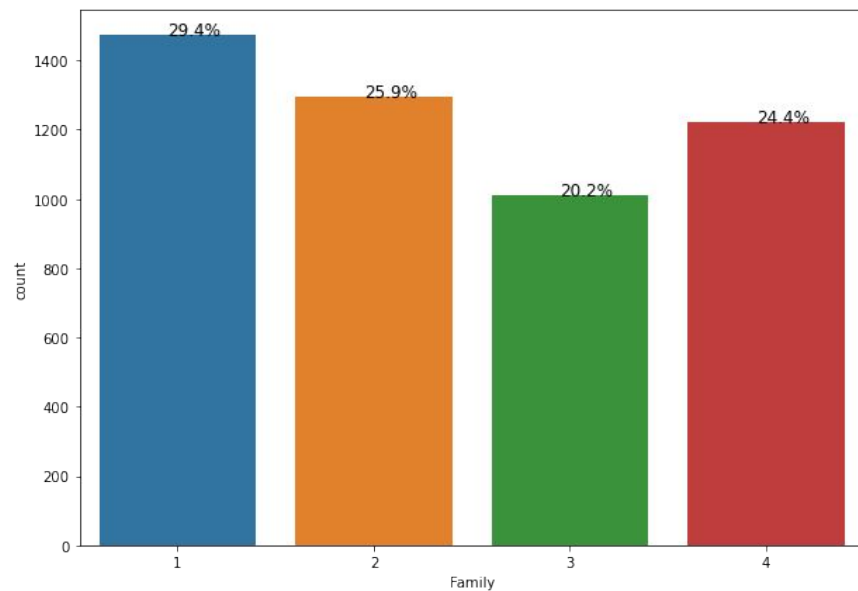


# EDA

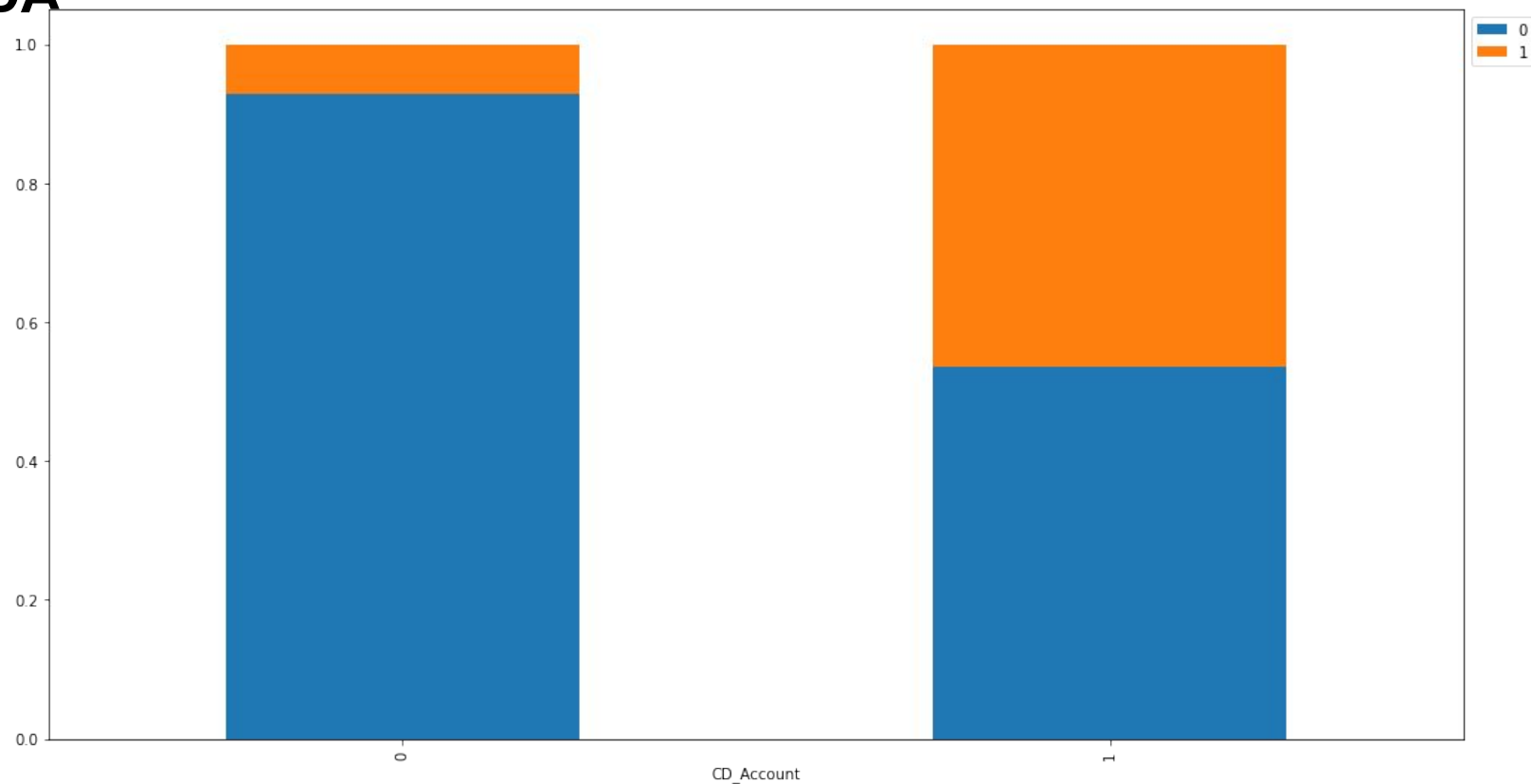




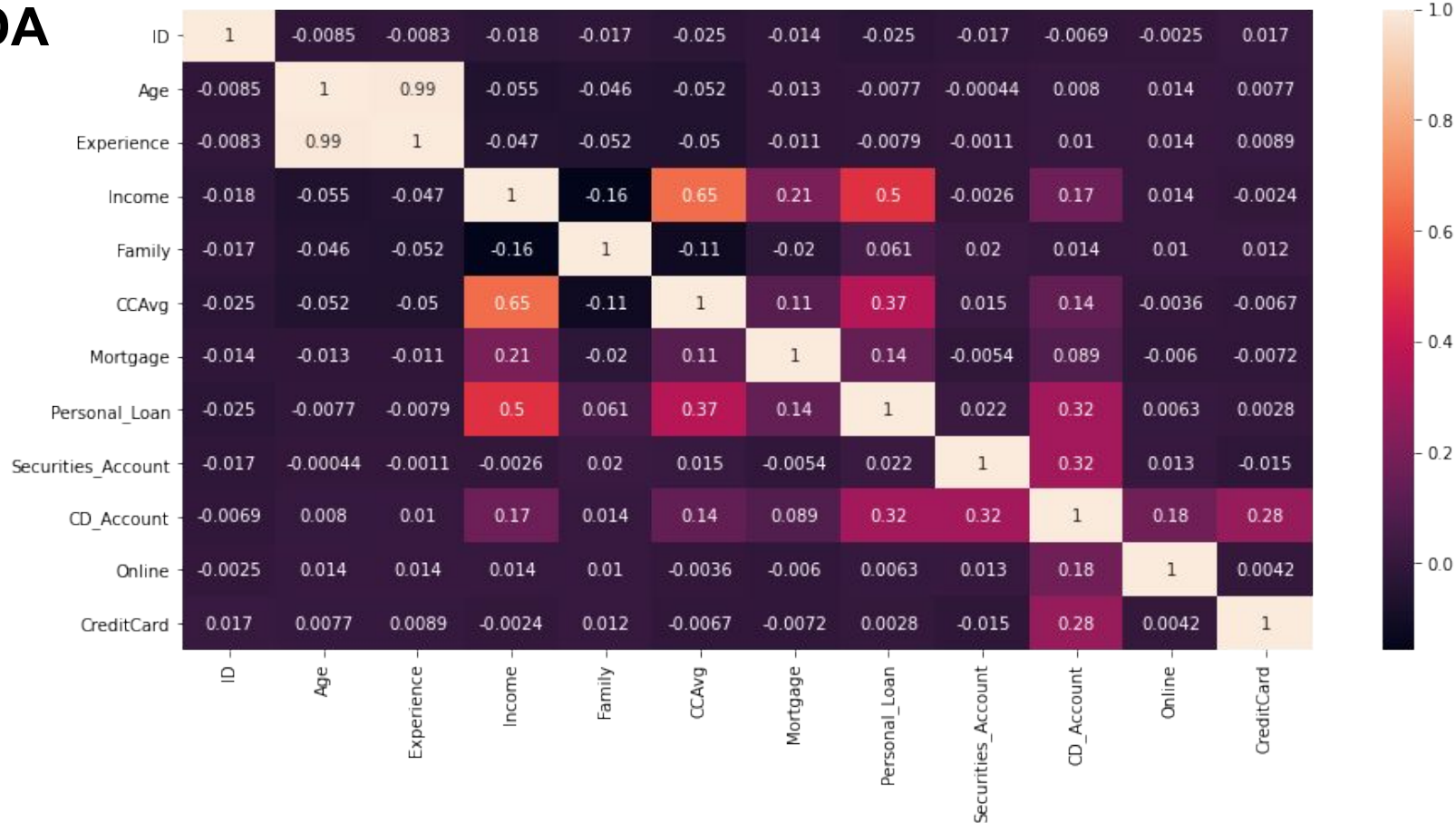
# EDA



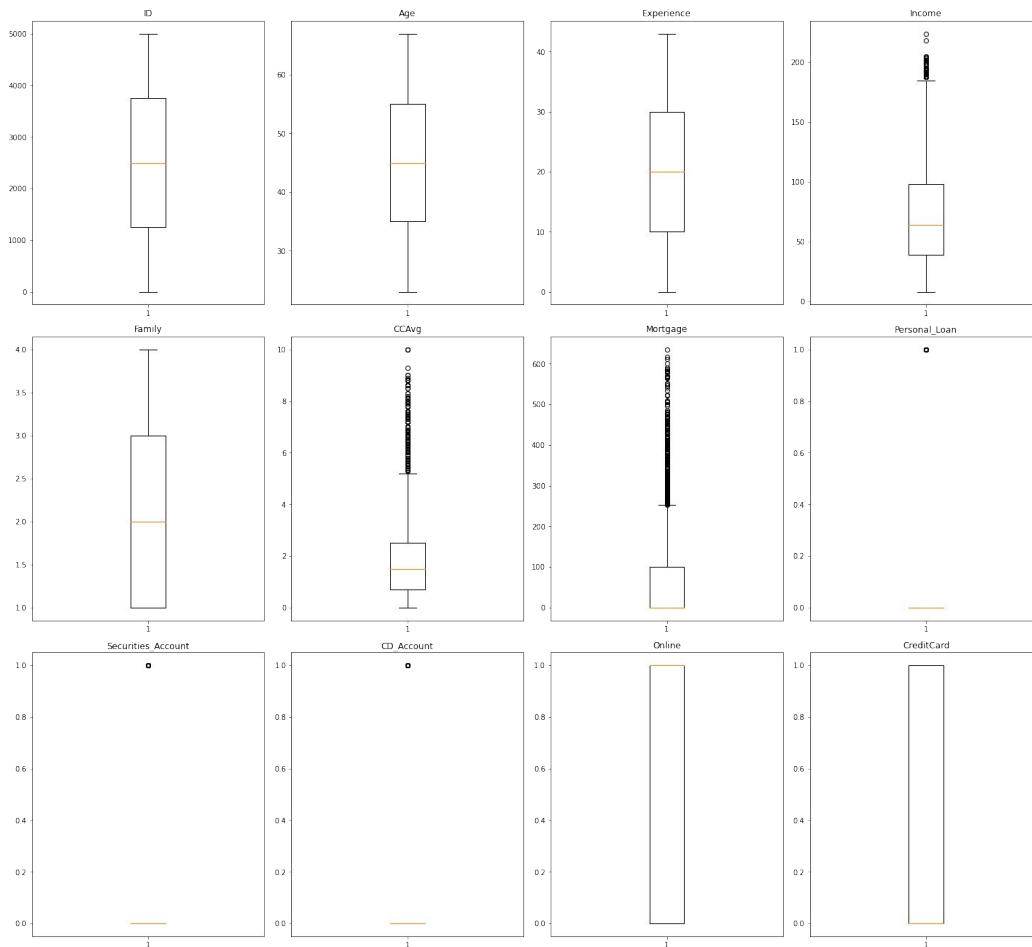
# EDA



# EDA



# EDA



# Logistic Regression Model Performance Summary

## Logit Regression Results

```
=====
Dep. Variable:    Personal_Loan  No. Observations:      3500
Model:           Logit  Df Residuals:      3482
Method:          MLE  Df Model:          17
Date:            Sat, 08 May 2021  Pseudo R-squ.:      0.6292
Time:            04:31:00  Log-Likelihood:     -412.06
converged:       True  LL-Null:         -1111.2
Covariance Type: nonrobust  LLR p-value:      3.645e-287
=====
```

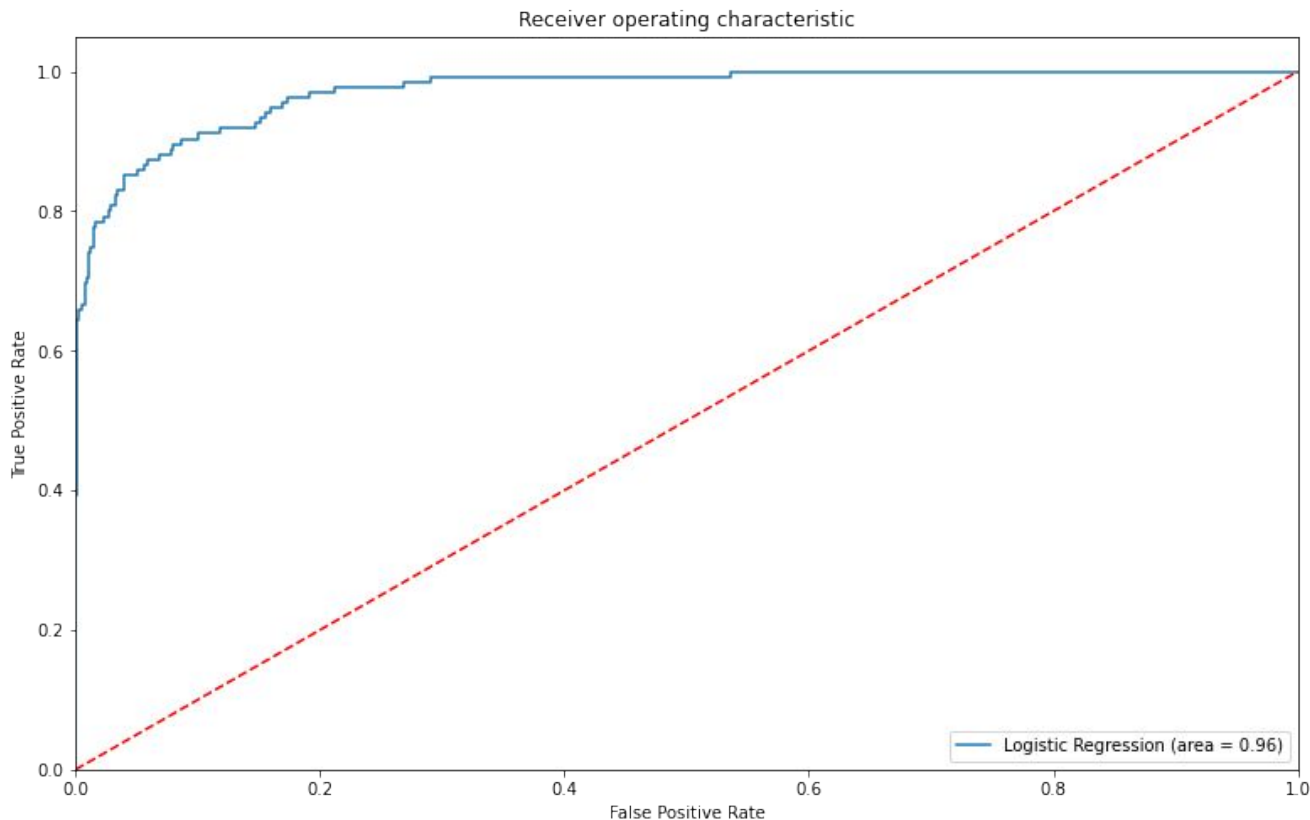
==

	coef	std err	z	P> z	[0.025	0.975]
const	-13.1444	0.739	-17.780	0.000	-14.593	-11.695
Experience	0.0066	0.008	0.808	0.419	-0.009	0.023
Income	0.0583	0.004	16.401	0.000	0.051	0.065
Family	0.5781	0.090	6.402	0.000	0.401	0.755
CCAvg	0.4374	0.069	6.381	0.000	0.303	0.572
Mortgage	0.0010	0.001	1.055	0.292	-0.001	0.003
Securities_Account	-0.8714	0.360	-2.420	0.016	-1.577	-0.166
CD_Account	3.7311	0.416	8.972	0.000	2.916	4.546
Online	-0.8127	0.200	-4.072	0.000	-1.204	-0.422
CreditCard	-1.1725	0.257	-4.568	0.000	-1.676	-0.669
Education_2	3.9897	0.320	12.474	0.000	3.363	4.617
Education_3	3.9409	0.309	12.754	0.000	3.335	4.546

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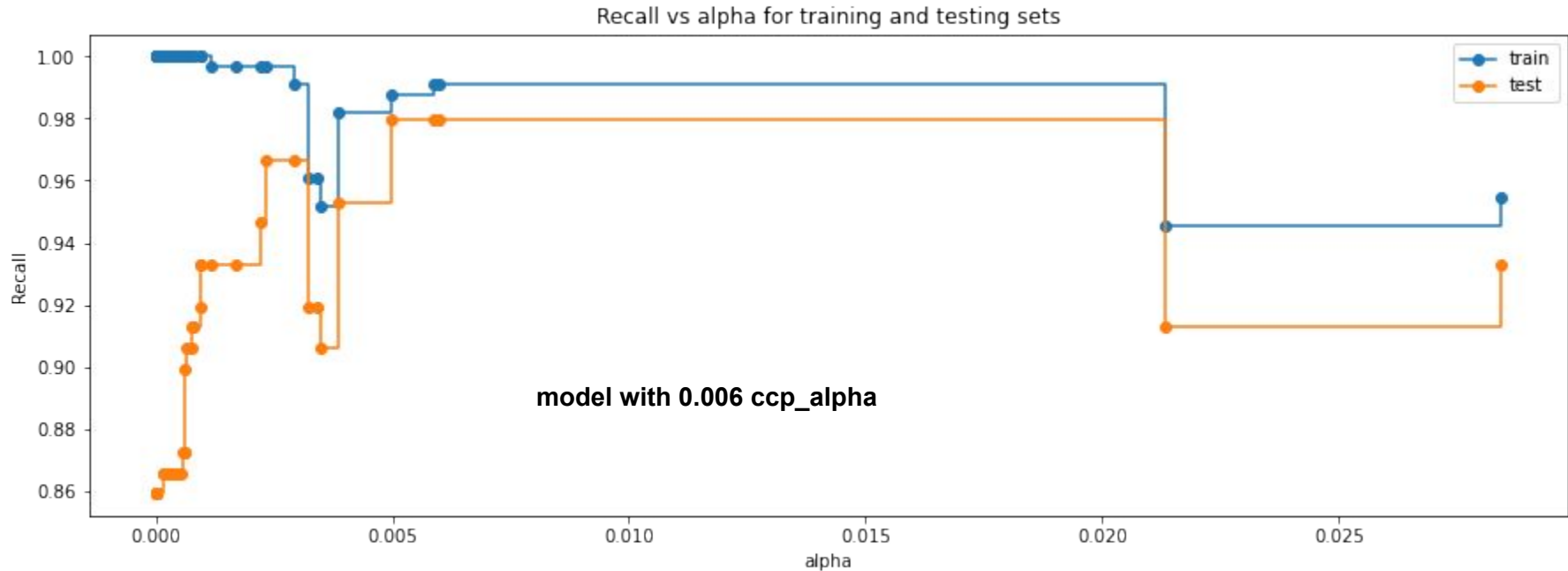
# Logistic Regression Model Performance Summary



# Logistic Regression Model Performance Summary

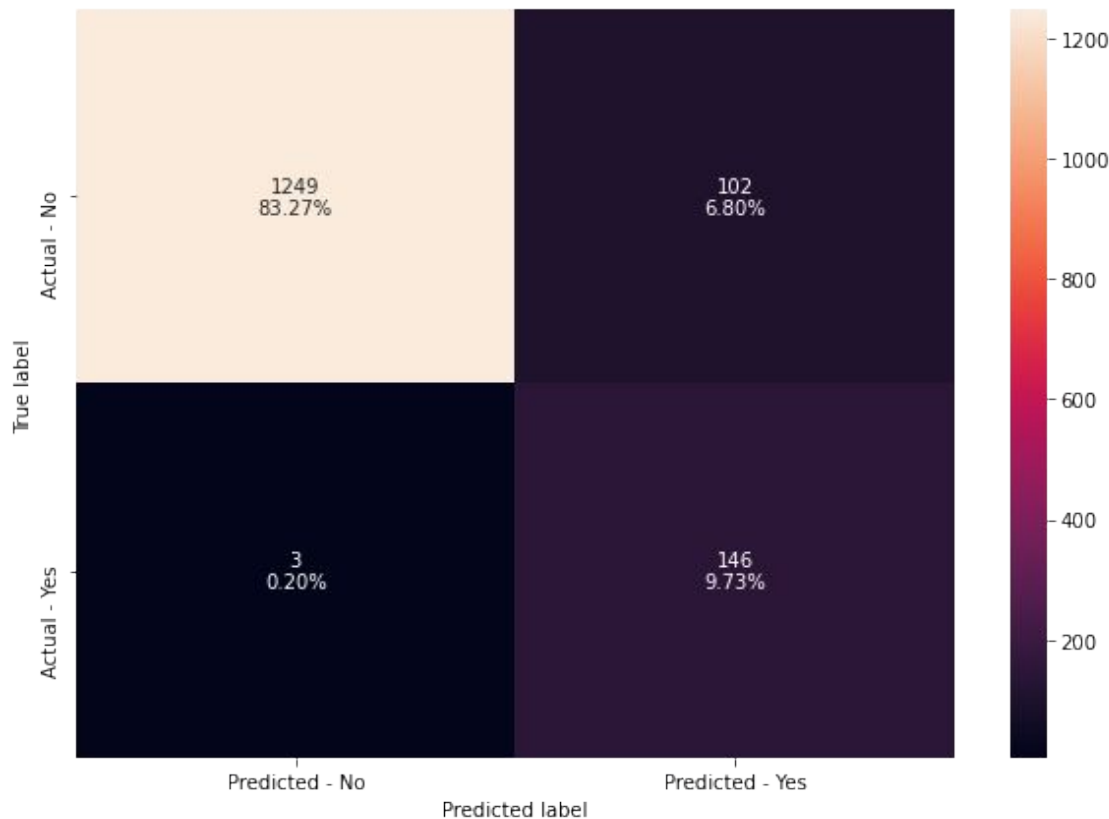
- Highest performing Logistic Regression model is : \*  
Logistic Regression with feature elimination
- Logistic Regression most important features are :
  - Education
  - CD\_Account

# Decision Tree Regression Model Performance Summary

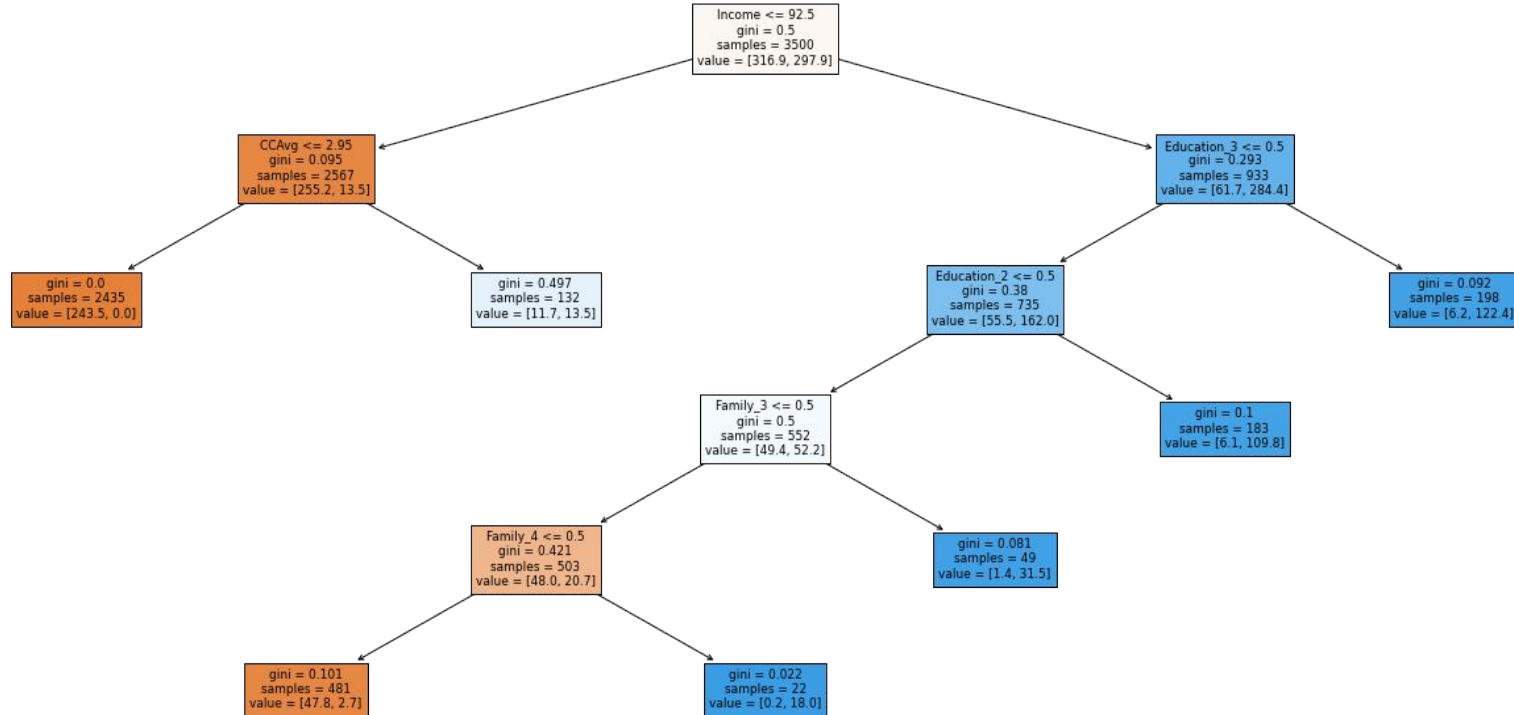




# Decision Tree Regression Model Performance Summary



# Decision Tree Regression Model Performance Summary



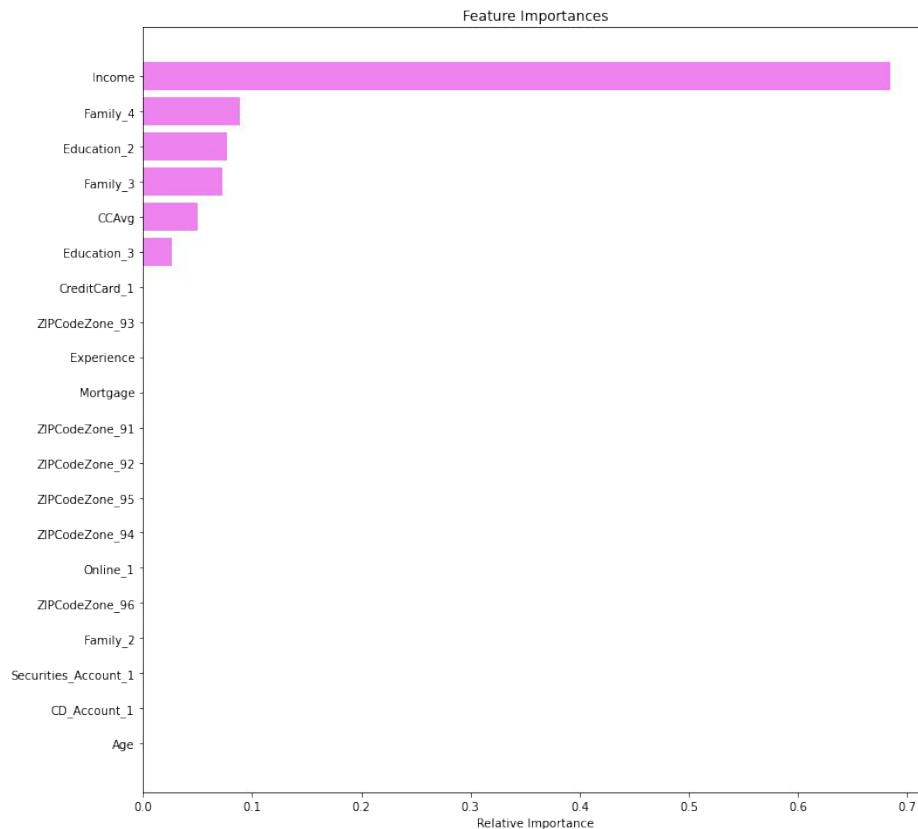
# Decision Tree Regression Model Performance Summary

```

|--- Income <= 92.50
| |--- CCavg <= 2.95
| | |--- weights: [243.50, 0.00] class: 0
| | |--- CCavg > 2.95
| | |--- weights: [11.70, 13.50] class: 1
|--- Income > 92.50
| |--- Education_3 <= 0.50
| | |--- Education_2 <= 0.50
| | | |--- Family_3 <= 0.50
| | | | |--- Family_4 <= 0.50
| | | | |--- weights: [47.80, 2.70] class: 0
| | | | |--- Family_4 > 0.50
| | | | |--- weights: [0.20, 18.00] class: 1
| | | |--- Family_3 > 0.50
| | | |--- weights: [1.40, 31.50] class: 1
| | |--- Education_2 > 0.50
| | |--- weights: [6.10, 109.80] class: 1
| |--- Education_3 > 0.50
| |--- weights: [6.20, 122.40] class: 1

```

# Decision Tree Regression Model Performance Summary



# Decision Tree Model Performance Summary

- Highest performing Decision Tree model is : Logistic Regression with feature elimination
- Decision Tree most important features are :
  - Income
  - Family\_4

# Conclusion

- Highest performing Logistic Regression model is :
  - **Logistic Regression with feature elimination**
- Logistic Regression most important features are :
  - Education
  - CD\_Account
- Highest performing Decision Tree model is :
  - **Decision Tree with Post-Pruning**
- Decision Tree most important features are :
  - Income
  - Family\_4

# Recommendations

- It is predicted that customers with higher education and that already have a CD\_Account are more likely to convert to Asset Customers.
- It is predicted that Customers with high Income and on average are a family of 4 are more likely to convert to Asset Customers.



# Happy Learning !



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Post Graduate Program in  
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PGP-DSBA-UTA-Dec20-A